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ON IGNORING THE SINGULARITY IN NUMERICAL QUADRATURE

BY

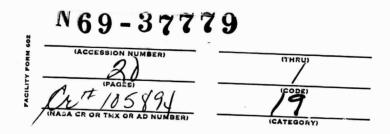
R. K. MILLER*

Center for Dynamical Systems

Division of Applied Mathematics

Brown University

Providence, Rhode Island



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On Ignoring the Singularity in Numerical Quadrature.

Ву

R.K. Miller

1. Introduction

Davis and Rabinswitz [1] recently studied the question of "ignoring the singularity" in numerical quadrature. That is, if f(t) becomes singular at a point ξ where $a \leq \xi \leq b$ then one defines $f(\xi) = 0$ (or any other finite value) and then approximates the integral.

$$I = \int_{a}^{b} f(t) dt$$

by a usual numerical quadrature rule. They show that this procedure is not valid in general. However if $\xi = a$ (or some other rational point of [a,b]), then compound quadrature rules do approximate I when f is monotone near ξ . Certain of the positive results in [1] where generalizied by Rabinswitz [2]. Gautschi [3] has applied a result in [2] to two other quadratures of interpolatory type.

The purpose of this paper is to generalize some of the convergence theorems in [1] and [2]. We shall replace the assumption of monotonicity of f(t) near $t=\xi$ by the more general condition that f(t) can be dominated by a monotone, integrable function. We shall also establish

some theorems on error bounds and convergence rates which are similar to those obtained in [4].

General Quadratures

Let M be the set

 $M = \{f \in C(0,T] \cap L^{1}(0,T) : f \text{ is nonnegative and non-increasing on } 0 < t \le T\}.$

Define M to be the set of all functions $f \in C(0,T]$ such that f can be majorizied by a function in M,

 $M_{d} = \{f \in C(0,T]: \exists F \in M \text{ with } |f(t)| \le F(t) \text{ on } 0 < t \le T\}.$

For any function $f \in M$ assign the arbitrary value f(t) = 0 when t = 0.

Given any numerical quadrature rule

$$Q(f) = \Sigma_{j=0}^{n} W(j) f(T(j)), f \in C[0,T]$$

let Q(f,S) be the modified quadrature rule obtained from Q by redefining f(0) = 0,

$$Q(f,S) = \sum_{j=0}^{n} W(j)f(T(j)), f(0) := 0$$

Then Q(f,S) ignores possible singularities at t=0 and is well defined for all functions $f\in M_d$. In general $Q(f,S)\neq Q$ (f) for all functions $f\in C[0,T]$. With these preliminaries we are ready to generalize the lemma [2]. First consider rules which are open at t=0.

Lemma 1. Consider a sequence of rules

$$Q_n(f) = \Sigma_{j=0}^n W_n(j) f(T_n(j))$$

where

$$0 < T_n (0) < T_n (1) < ... < T_n(m_n) \le T$$

and $W_n(j)>0$ for all j. Define $T_n(-1)=0$.

Suppose there exist positive constants C and A such that uniformly for all positive integers n, if $j = O(1)m_n$ and if $|T_n(j)| < A$ then

(2.1)
$$W_n(j) \leq C \{T_n(j) - T_n(j-1)\}.$$

$$(j = 0(1)m_n = 0,1,2,...,m_n)$$

Suppose for each function $g \in C[0,T]$ one has

(2-2)
$$\lim Q_n(g) = \int_0^T g(t)dt, n\to\infty.$$

Then for any function $f \in M_d$

(2-3)
$$\lim_{n \to \infty} Q_n(f,S) = \int_0^T f(t)dt, n \to \infty.$$

In particular if O<B<A, if one defines

$$f_B(t) = f(t) \underline{on} B \leq t \leq T; = f(B) \text{ or } 0 \leq t \leq B,$$

and if

$$\delta(t) = \sup \{ |f(s) - f_B(s)| : t \le s \le T \},$$

then the error

$$E_s(f,Q_n) = \int_0^T f(t)dt - Q_n(f,S)$$

satisfies the estimate

$$|E_{s}(f,Q_{n})| \leq |E(f_{B},Q_{n})| + |f_{0}^{B}\{f(t) - f(B)\}dt| + cf_{0}^{B} \delta(t)dt.$$

Proof. Write $E = E_s(f,Q_n)$ in the form

$$\begin{split} \mathbf{E} &= \int_{0}^{\mathbf{T}} \{ \mathbf{f}(\mathbf{t}) - \mathbf{f}_{\mathbf{B}}(\mathbf{t}) \} d\mathbf{t} + \mathbf{E}(\mathbf{f}_{\mathbf{B}}, \mathbf{Q}_{\mathbf{n}}) + \mathbf{\Sigma}_{\mathbf{j}=0}^{\mathbf{m}_{\mathbf{n}}} \mathbf{W}_{\mathbf{n}}(\mathbf{j}) \\ & \{ \mathbf{f}_{\mathbf{B}}(\mathbf{T}_{\mathbf{n}}(\mathbf{j})) - \mathbf{f}(\mathbf{T}_{\mathbf{n}}(\mathbf{j}) \} \\ &= \int_{0}^{\mathbf{B}} \{ \mathbf{f}(\mathbf{t}) - \mathbf{f}(\mathbf{B}) \} d\mathbf{t} + \mathbf{E}(\mathbf{f}_{\mathbf{B}}, \mathbf{Q}_{\mathbf{n}}) + \mathbf{\varepsilon}_{\mathbf{n}}. \end{split}$$

Then for any n

$$\begin{aligned} \left| \varepsilon_{\mathbf{n}} \right| &\leq \Sigma_{\mathbf{j}=0}^{m} \ \mathbb{W}_{\mathbf{n}}(\mathbf{j}) \ \left| \mathbf{f}(\mathbb{T}_{\mathbf{n}}(\mathbf{j})) - \mathbf{f}_{\mathbf{B}}(\mathbb{T}_{\mathbf{n}}(\mathbf{j})) \right| \\ &\leq \Sigma_{\mathbf{j}=0}^{m} \ \mathbb{W}_{\mathbf{n}}(\mathbf{j}) \ \delta \ \left(\mathbb{T}_{\mathbf{n}}(\mathbf{j})\right) = \mathbb{Q}_{\mathbf{n}}(\delta, \mathbf{S}) \end{aligned}$$

Since $f \in M_d$, there exists a majorizing function $F \in M$. Then for s in the range $b < t \le s < B$ one has

$$|f(s) - f_B(s)| \le F(s) + F(B) \le 2F(t)$$
.

Therefore

$$\delta$$
 (t) $\leq 2F(t)$ on 0 < t \leq B, $\delta(t) = 0$

on B \leq t \leq l, and hence $\delta \in L^1(0,1)$. Note also that $\delta(t)$ is continuous, nonnegative and nonincreasing. This

together with (2.1) implies that

$$\begin{split} Q_n(\delta,S) &= \Sigma_{T_n(j) < B} \ W_n(j) \delta(T_n(j)) \leq C \Sigma \{T_n(j) - T_n(j-1)\} \delta(T_n(j)) \leq C \ \Sigma \ \int_{T_n(j-1)}^{T_n(j)} \delta \\ &\qquad \qquad (t) dt = C \ \int_0^B \delta \ (t) dt \,. \end{split}$$

This completes the proof of (2.4).

Line (2.3) follows immediately from (2.4) and the estimate $\delta(t) \leq 2$ F(t). Indeed by first choosing B small and then choosing n large one can make the right hand side of (2.4) as small as desired. Q.E.D. Almost the same result is true for quadratures which are closed at t=0.

Lemma 2. Suppose a sequence of quadrature rules Q_n satisfies the two conditions

$$0 = T_n(0) < T_n(1) < ... < T_n(m_n) \le T, W_n(j) > 0.$$

Suppose there exist positive constants C and A such that $\frac{\text{if }|T_n(j)| < A \text{ and if } j = 1(1)m_n \text{ then } (2.1) \text{ is true}}{\text{uniformly in n. If } (2.2) \text{ is also true then } (2.3) \text{ follows.}}$ In particular if $f_b \in C[0,T]$ and $\delta \in M_d$ are the functions defined in Lemma 1 and if 0 < B < A, then

(2.5)
$$|E_s(f,Q_n)| \le |E(f_B,Q_n)| + |f_0^B(f(t) - f(B))dt| + C |f_0^B(f(t))dt| + W_n(0) |f(B)|.$$

<u>Proof.</u> The proof is the same as that of Lemma 1 except for the estimates of ϵ_n . In this case

$$\varepsilon_{n} = \Sigma_{j=0}^{m} W_{n}(j) \{f(T_{n}(j)) - f_{E}(T_{n}(j))\},$$

and

$$|\varepsilon_n| \le Q_n(\delta,S) + W_n(0) |f(B)|$$

 $\le C \int_0^B \delta(t) dt + W_n(0) |f(B)|$

where we define $\delta(0)=0$. Our hypotheses easily imply that $W_n(0) \to 0$ as $n \to \infty$. Therefore Lemma 2 is proved. Combining the two results we have proved: Theorem 1. Consider a sequence of numerical quadrature rules Q_n where

$$0 \le T_n(0) < T_n(1) < ... < T_n(m_n) \le T_{n}(j) > 0.$$

Suppose (2.1) is true as in Lemma 1 (when $T_n(0)>0$)

or Lemma 2 ($T_n(0)=0$). If (2.2) is also true, then for any $f \in M_d$,

$$Q_n(f) \rightarrow \int_0^T f(t)dt \text{ as } n \rightarrow \infty.$$

Indeed if 0 < B < A and if $f_B \in C$ [0,T] and $\delta \in M_d$ are the functions defined in Lemma 1 then

(2-6)
$$|E_s(f,Q_n)| \le |E(f_B,Q_n)| + |f_0^B\{f(t) - f(B)\}dt| + Cf_0^B \delta(t)dt + \{1 - sgn T_n(0)\}$$

$$W_n(0)|f(B)|.$$

One can also generalize the Corollary in [2, p.196] in the obvious way.

3. Compound Rules.

Consider a quadrature rule R defined on the interval $0\,\leq\,t\,\leq\,1\,:$

(R)
$$R(t) = \sum_{j=0}^{J} W_j f(t_j)$$

where J > 0 and

(3-1)
$$0 \le t_0 < t_1 < \dots < t_{J-1}, W_j > 0, \Sigma_{j=0}^m W_j = 1.$$

(If $t_0 > 0$ then define $t_{-1} = 0$.) For any integer $n \ge 1$

and any interval $0 \le t \le T$ one can then define a compound rule

$$(n \times R)$$
 $R_n(f) = \sum_{k=0}^{n-1} \{\sum_{j=0}^{J} HW_j f(t_j H + Hk)\}$

where H = T/n. Let C>O be any constant satisfying

(3.2)
$$W_{j} \le (t_{j} - t_{j-1}) c$$
 $(j = l(1)J)$

and either

(3.3a)
$$W_0 \le (t_0 + 1 - t_J) C$$
 (if $t_0 > 0$ or $t_J < 1$)

or

(3.3b)
$$(W_0 + W_J) \leq (t_J - t_{J-1}) C$$
 (if $t_0 = 0$ and $t_J = 1$).

Theorem 2. If (R) satisfies (3.1) then for any $f \in M_d$

$$\lim_{n\to\infty} R_n(f,S) = f_0^T f(t) dt.$$

<u>Proof.</u> The definition (3.2-3) of C implies that (2.1) is true with A=T. Since R integrates constants and $n\to\infty$ then (2.2) is also trival. Therefore Theorem 2 is a corollary of Theorem 1. Q.E.D.

The error estimate (2.6) is rather pessimistic for compound rules. Therefore we shall derive another estimate which is more suitable for many purposes. Let K>O be the smallest constant which satisfies

$$(3.4)$$
 $W_{j} \leq (t_{j} - t_{j-1}) K$

for j = 1(1)J if $t_0=0$ or for j=0(1)J if $t_0>0$. Theorem 3. Suppose (3.1) and (3.4) are satisfied. Let H = T/n for any function $f \in M_d$ define

$$f_H(t) = f(H) \underline{if} 0 \le t \le H; f(t) \underline{if} H \le t \le T.$$

If $F \in M$ is any majorizing function for f then

(3.5)
$$|E_s(f,R_n)| \le |E(f_H,R_n)| + \int_0^H F(t)dt + K \int_0^H F(t)dt$$

or

$$|E_{S}(f,R_{m})| \leq |E(f_{H},R_{n})| + (1+K) \int_{0}^{H} F(t)dt.$$

<u>Proof.</u> Since (R) integrates constants, then the error may be written in the form

$$E_s(f,R_n) = E(f_H,R_n) + E_0$$

where

(3.6)
$$E_0 = \int_0^H \int_0^H f(t) dt - \sum_{j=0}^J Hw_j f(t, H).$$

Since $|f(t)| \le F(t)$ on $0 < t \le T$, then

$$\left|\int_0^H f(t)dt\right| \leq \int_0^H F(t)dt$$
.

Let α = 0 if t_0 > 0 and α = 1 if t_0 =0. Then f(0) = 0 and (3.4) imply

$$\begin{split} \left|\Sigma_{\mathbf{j}=0}^{J} & \text{Hw}_{\mathbf{j}} & \text{f(t,H)} \right| \leq \Sigma_{\alpha}^{J} & \text{Hw}_{\mathbf{j}} & \text{F(t,H)} \\ \\ & \leq K \sum_{\alpha}^{J} & \text{H(t}_{\mathbf{j}} - \mathbf{t}_{\mathbf{j}-1}) & \text{F(t,H)} \\ \\ & \leq \sum_{\alpha}^{J} \int_{\mathbf{t}_{\mathbf{j}-1}}^{\mathbf{t}_{\mathbf{j}}} & \text{F(t)} & \text{dt} = K \int_{0}^{\mathbf{t}} \mathbf{J}^{H} & \text{F(t)} & \text{dt}. \end{split}$$

This proves (3.5) and the theorem.

If one knows that $f \in C^{1}(0,T]$, then the term $E(f_{H},R_{n})$ may be estimated using Peano's theorem.

That is

$$E(f_{H},R_{n}) = \int_{0}^{1} P_{n}(s) f'_{m}(s) dt$$

where P_n is the appropriate Peano kernal. Since

 $P_n(s+H) = P_n(s)$ on $0 \le s \le T-H$, then $|P_n(s)|$ need be be estimated only on the interval 0 < s < H. Therefore the following result is an immediate corollary of Theorem 3.

Corollary 1. Assume the hypotheses of Theorem 3. If $f \in C^1(0,T]$ then

(3.7)
$$|E(f,R_n)| \leq |P_n| |\int_H^T |f'(t)| dt + \int_0^H F(t) dt + K \int_0^T f'(t) dt$$

where $||P_n|| = \sup \{|P_n(s)| = 0 < s < H\}$. Corollary 1 is useful in estimating convergence rates in certain cases. Following [4] we shall say that a function $f \in C^1(0,T]$ is weakly singular at t=0 if the function

$$\alpha(t,f) = |f(t)| + \int_{t}^{T} |f'(s)| ds$$

is in $L^1(0,T)$.

Corollary 2. Suppose the hypotheses of Theorem 3 are true.

If f is weakly singular (at t = 0) then

(3.8)
$$E_s(f,R_n) = O(\int_0^H \alpha(t,f)dt) \underline{as} H \rightarrow 0.$$

<u>Proof.</u> It follows immediately from the definition of weakly singular functions that $f \in M_d$ and $\alpha(t,f) \in M$ is a majorizing function. Thus (3.7) implies

$$\begin{split} |E_{s}(f,R_{n})| & \leq ||P_{n}|| \int_{H}^{T} |f'(t)dt + \int_{0}^{H} \alpha(t,f)dt + \\ & \qquad K \int_{0}^{H} \alpha(t,f)dt \\ & \leq ||P_{n}|| \alpha(H,f) + (1+K) \int_{0}^{H} \alpha(t,f)dt. \end{split}$$

Using the estimate $||P_n|| \le 2H$ (see for example [4, section II]) and the monotonicity of α one has

$$|P_n|$$
 $\alpha(H,f) \leq 2H \alpha(H,f) \leq 2\int_0^H \alpha(t,f) dt$.

Therefore for any n (H = T/n) one has

$$|E(f,R_n)| \le (3 + K) \int_0^H \alpha(t,f)dt.$$
Q.E.D.

For example if $f(t) = t^{-P}$ (0<p<1) then (3.8) predicts that $E_s(f,R_n)$ is at least of order h^{I-P} . If $f(t) = t^{-P}$ sin (t^{-q}) where 0 <p,q <1 and p + q <1 then $E_s(f,R_n)$ is at least of order h^{I-p-q} . If $f(t) = t^{-P}$ sin (t^{-q}) where 0 0 and P + q \geq 1 then our theory predicts convergence but gives no order estimate.

4. Numerical Example.

The data in [1] will be used to illustrate the theory given above. For the midpoint rule $M(f) = f(\frac{1}{2})$ one has H = h = T/n. Since $-P_n(s) = s$ if 0 < s < h/2; 2-h if h/2 < s < h, then $||P_n|| = h/2$ and K = 2. Therefore (3.2) has the form

$$|E_s(f,M_n)| \le (h/2) \int_n^T |f'(t)| dt + \int_0^h F(t) dt + 2 \int_0^{h/2} F(t) dt.$$

Table 1 contains data for the case

(4.1)
$$\int_0^1 t^{-1/2} dt = 2$$
.

The fourth column is the theoretical error computed using (3.7). This error bound is seen to be pessimistic by a factor of 7 to 8. Corollary 2 suggests that the error may be of the approximate form

$$(4.2) \quad \mathbb{E}_{s}(f,M_{n}) = C\sqrt{h} \quad (h=T/n)$$

for some constant C > 0. The ratios $E_s(f,M_n)$ / $E_s(f,M_{n+1})$ are given in column five. The theoretical ration computed using (4.2) is $\sqrt{2}$ (column six). It can be seen that (4.2) is approximately true with C = .61.

(Insert table 1 near here)

Table 2 contains similar data for (4.1) using the trapezoid rule (T), Simpson's rule (S) and the Gaussian two point rule (G_2). The theoretical errors are good for the trapezoid rule and progressive worse for Simpson and Gauss two point. In all cases (4.2) is approximately true, $E_s(f,T_n) = 1.5\sqrt{h}$, $E_s(f,S_n) = .89\sqrt{h}$ and $E_s(f,nXG_2) = .35\sqrt{h}$.

(Insert table 2 near here).

One could also analysis the data in [1] for (4.1) using the Gauss 48 - point rule. In this case $E_s(f,n \times G_{48})$ = $.18\sqrt{h}$ (approximately). It would be very difficult to estimate $||P_n||$ accurately in this case. Thus (3.7) is essentially useless.

Data for the example

$$f_0^1 t^{-\frac{1}{2}} \sin(t^{-\frac{1}{2}}) dt = 1.08134$$

is given in table 3 for Simpson's rule and for the Gauss 48 - point rule. For this case our theory predicts convergence but yeilds no useful information on errors or convergence rates. For Simpson's rule the method appears to be convergent. For G_{48} the method starts to converge nicely but blows up at the fourth step.

TABLE 1

| n . | $M_{n}(f_{0})$ | Error | Th.Error | Ratio | Th.Ratio |
|-----------------|----------------|-------|----------|-------|----------|
| 25 | 1.8931 | .1069 | .6763 | 1.414 | 1.414 |
| 26 | 1.9244 | .0756 | .4815 | 1.413 | 1.414 |
| 27 | 1.9465 | .0535 | .4321 | 1.415 | " |
| 28 | 1.9622 | .0378 | .2427 | 1.416 | n |
| 29 | 1.9733 | .0267 | .1720 | 1.413 | " |
| 2 ¹⁰ | 1.9811 | .0184 | .1218 | 1.410 | " |
| 2 ¹¹ | 1.9866 | .0134 | .0864 | | |

TABLE 2

| 2 ⁵ x T | Approx. | Error | Th.Error | Ratio | Th.Ratio |
|---------------------|-----------------|---------------|----------|-------|----------|
| | 1.7418 | .2582 | .6031 | 1.414 | 1.414 |
| 2 ⁶ хт | 1.8174 | .1826 | .4294 | 1.414 | ** |
| 2 ⁷ X T | 1.8709 | .1291 | .3055 | 1.414 | " |
| 2 ⁸ X T | 1.9087 | .0913 | .2168 | 1.415 | 11 |
| 2 ⁹ X T | 1.9355 | .0645 | .1537 | 1.414 | TI . |
| 2 ¹⁰ X T | 1.9544 | .0456 | .1089 | | |
| | | | | | |
| 2 ⁵ x s | 1.8427 | .15 73 | 1.1792 | 1.413 | 1.414 |
| 2 ⁶ x s | 1.8887 | .1113 | .8735 | 1.414 | " |
| 2 ⁷ x s | 1.9213 | .0787 | .6198 | 1.415 | 11 |
| 2 ⁸ x s | 1.9444 | .0556 | .4393 | 1.411 | " |
| 2 ⁹ x s | 1.9606 | .0394 | 3113 | 1.412 | " |
| 2 ¹⁰ x s | 1.9721 | .0297 | .2203 | | |
| | | | | | |
| 2 X G2 | 1.7528 | .2472 | 7.6572 | 1.414 | 1.414 |
| 4 X G ₂ | 1.8252 | .1748 | 5.4012 | 1.225 | 1.225 |
| 6 x G ₂ | 1.8573 | .1427 | 4.4213 | 1.155 | 1.155 |
| 8 X G ₂ | 1.8764 | .1236 | 3.8341 | 1.118 | 1.118 |
| 10 X G ₂ | 1.8894 | .1106 | 3.4328 | 1.096 | 1.095 |
| 12 X G ₂ | 1.8 9 91 | .1009 | 3.1360 | 1.080 | 1.080 |
| 14 X G ₂ | 1.9066 | .0934 | 2.9050 | 1.069 | 1.069 |
| 16 X G ₂ | 1.9126 | .0874 | 1.7640 | | |
| | | | | | |

TABLE 3

| 2 ⁵ x s 2 ⁶ x s | Approx 1.1234 .9116 | Error 0421 .1697 | Ratio .241 1.562 |
|--|---------------------------|------------------------|------------------------|
| 2 ⁷ x s 2 ⁸ x s | .9727 | .1086 | 1.017 |
| 2 ⁹ x s | 1.0201 | .0612 | 2.696 |
| | .9449 | .1346 | 1.534 |
| 1 x G ₄₈ 2 x G ₄₈ | .9924 | .0889 | 2.285 |
| 3 X G ₄₈ | 1.0402 .9300 | .0389 .1513 | .2571 |

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